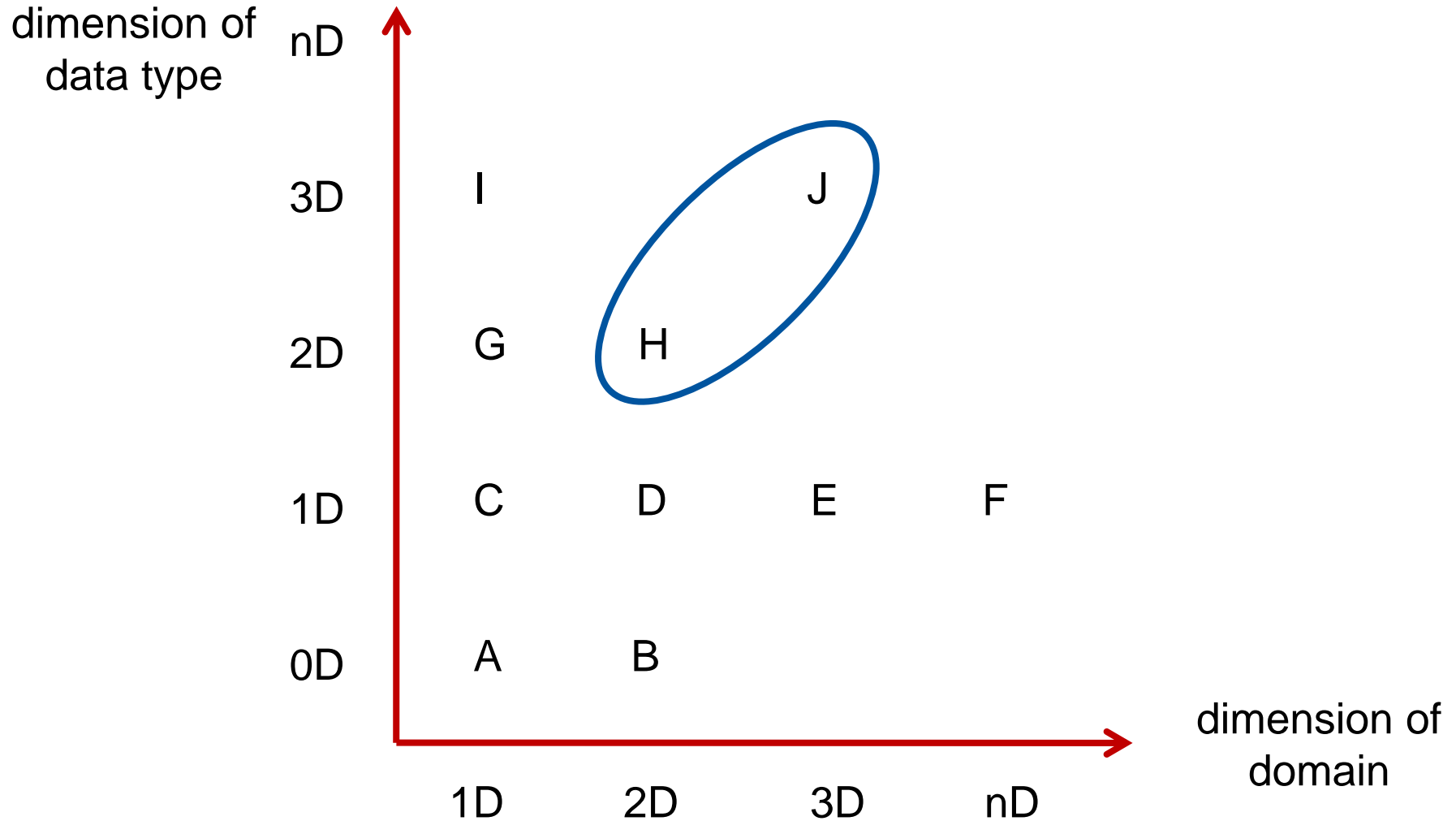


Vector Field Visualization

Leif Kobbelt

Types of Data



Characteristic Lines

- Types of characteristic lines in a vector field:
 - stream lines: tangential to the vector field
 - path lines: trajectories of massless particles in the flow (non-static flow fields)
 - streak lines: trace of dye that is deposited into the flow at a fixed position
 - time lines (time surfaces): propagation of a line (surface) of massless elements in time

Characteristic Lines

- stream lines
 - tangential to the vector field
 - stationary vector field (arbitrary, yet fixed time t)
 - stream line is a solution to the initial value problem of an ordinary differential equation:

$$\mathbf{L}(0) = \mathbf{x}_0 \quad , \quad \frac{d\mathbf{L}(u)}{du} = \mathbf{v}(\mathbf{L}(u))$$

initial value
(seed point \mathbf{x}_0)

ordinary differential equation

Characteristic Lines

- path lines
 - trajectories of massless particles in the flow
 - vector field can be time-dependent (unsteady)
 - path line is a solution to the initial value problem of an ordinary differential equation:

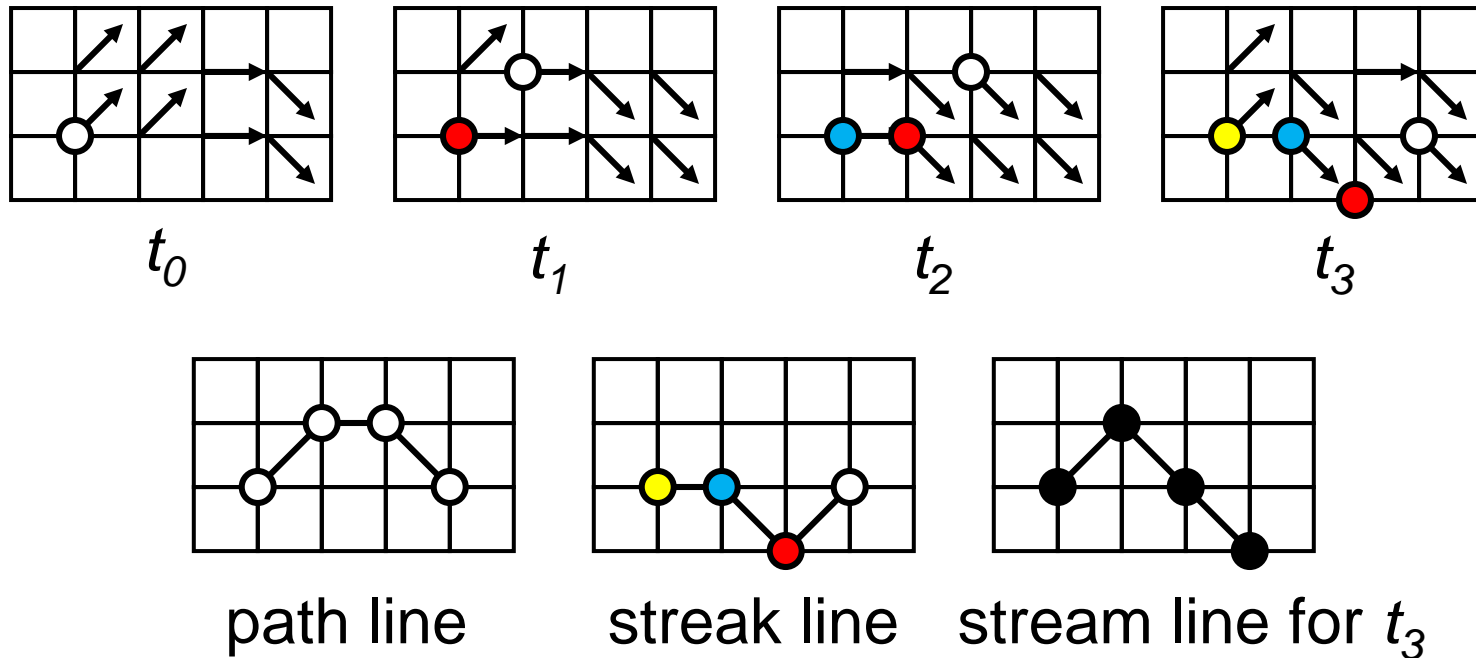
$$\mathbf{L}(0) = \mathbf{x}_0 \quad , \quad \frac{d\mathbf{L}(t)}{dt} = \mathbf{v}(\mathbf{L}(t), t)$$

Characteristic Lines

- streak lines
 - trace of dye that is released into the flow at a fixed position
 - connect all particles that passed through a certain position (non-stationary flow)
- time lines (time surfaces)
 - propagation of a line (surface) of massless elements over time
 - many point-like particles that are traced synchronously
 - connect particles that were deposited simultaneously

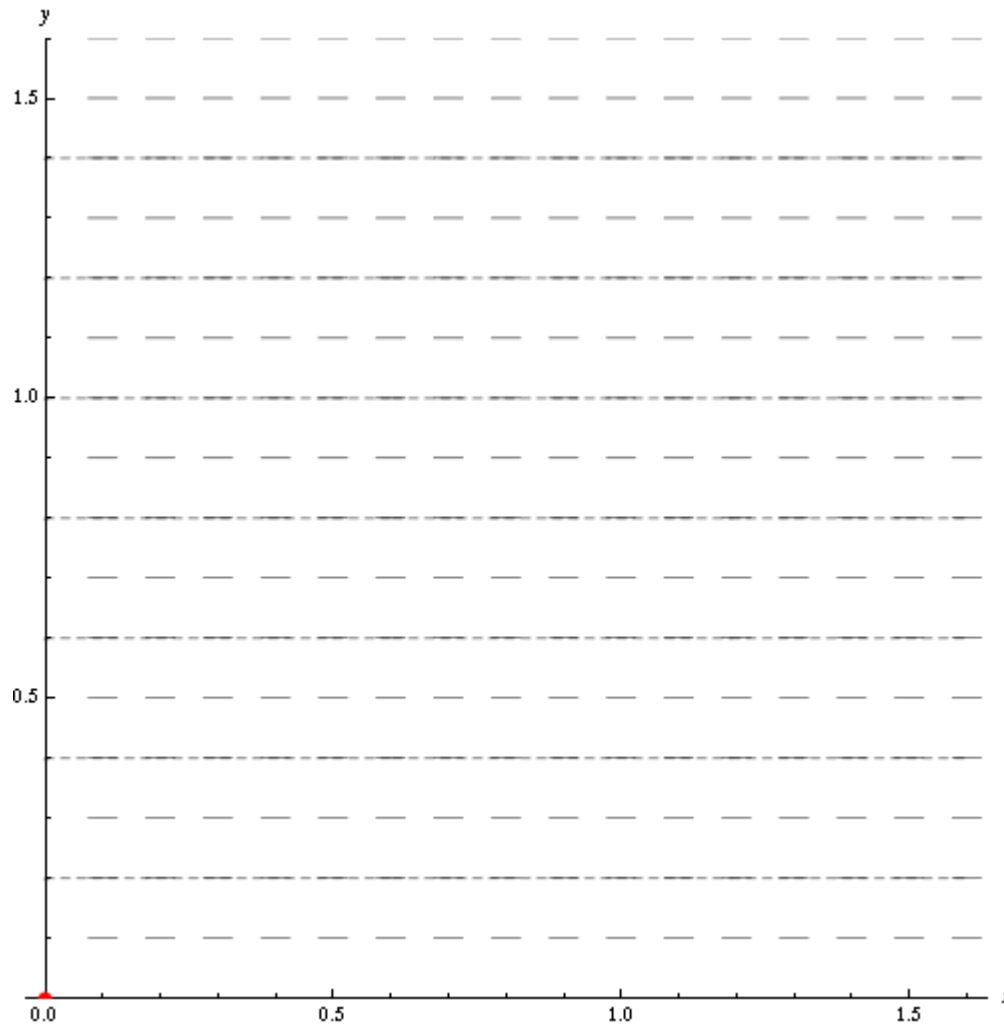
Characteristic Lines

- comparison of path lines, streak lines, and stream lines



- path lines, streak lines, and stream lines are identical for stationary flows

Characteristic Lines



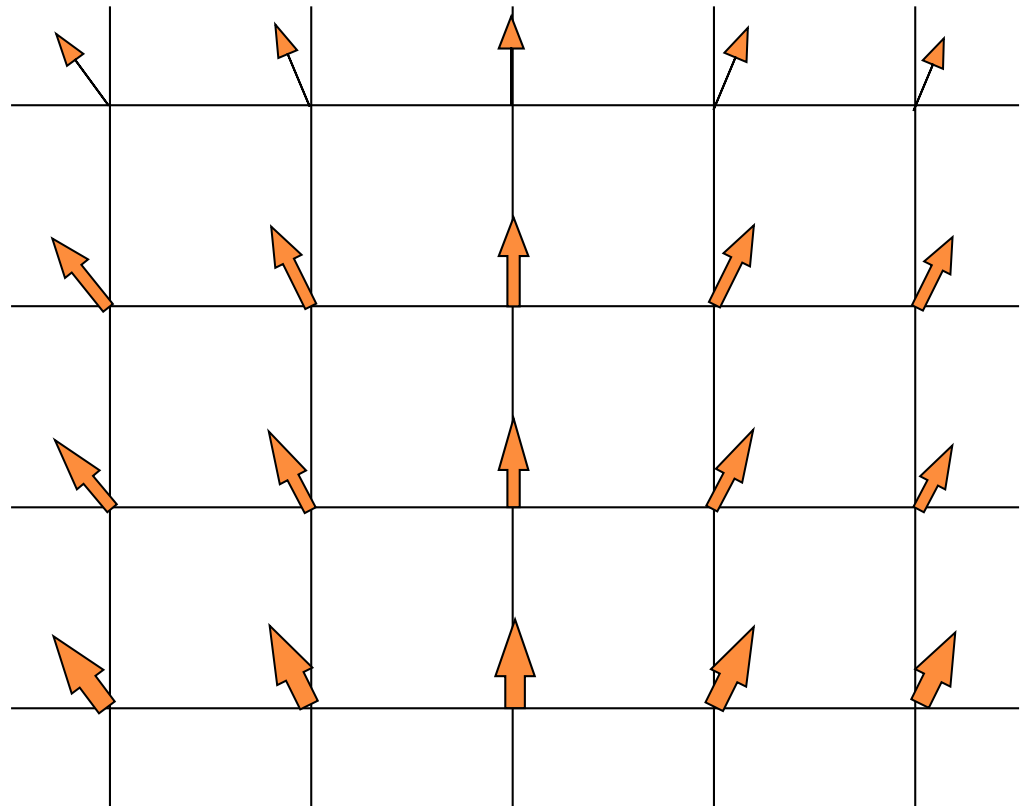
Arrows and Glyphs

- visualize **local** features of the vector field:
 - vector itself
 - vorticity
 - external data: temperature, pressure, etc.
- important elements of a vector:
 - direction
 - magnitude
 - not: components of a vector
- approaches:
 - arrow plots
 - glyphs

Arrows and Glyphs

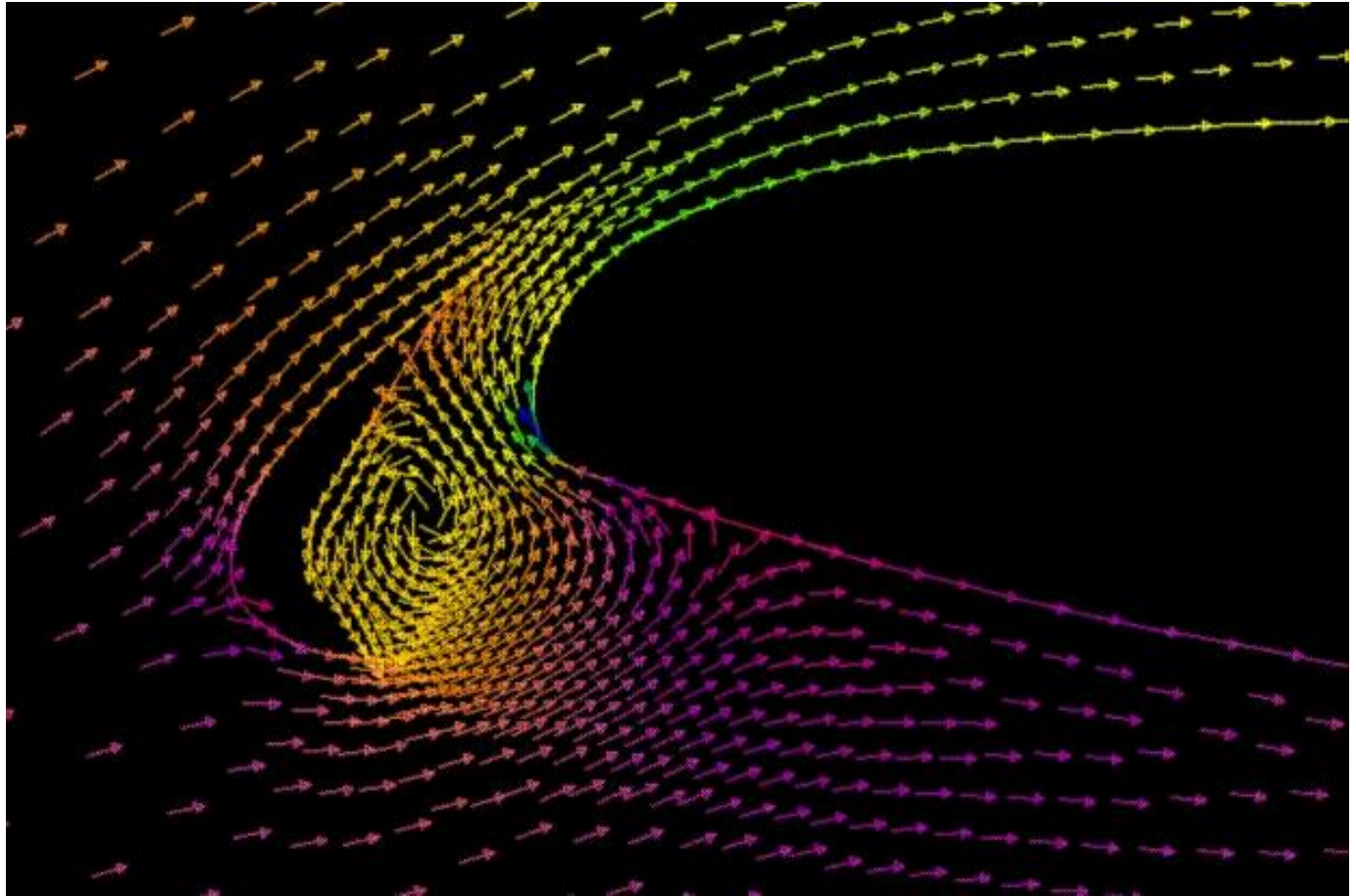
- arrows visualization

- direction of vector field
- orientation
- magnitude:
 - length of arrows
 - color coding



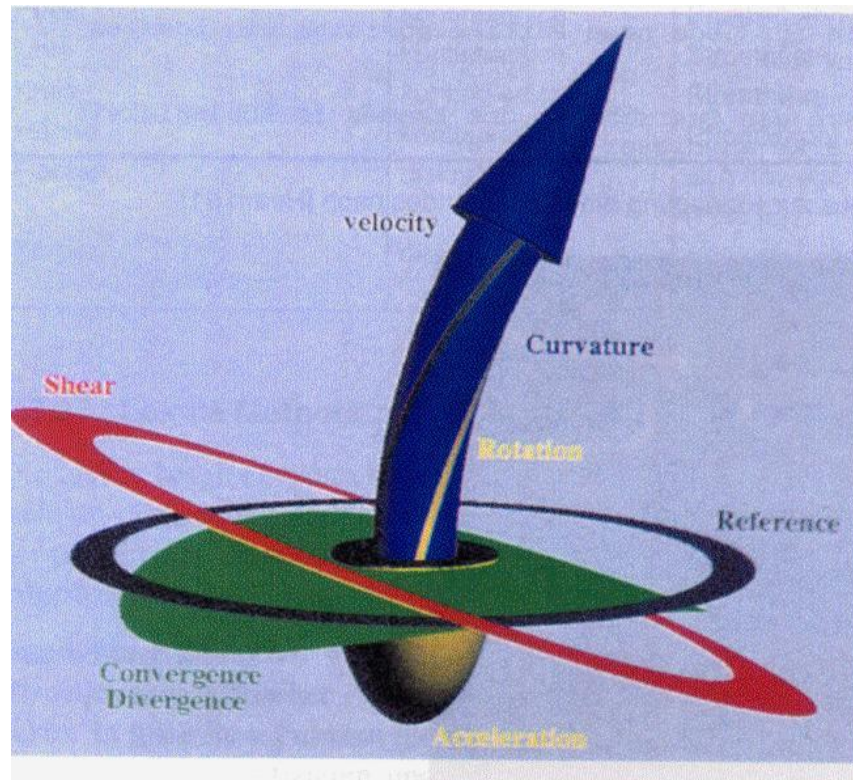
Arrows and Glyphs

- arrows



Arrows and Glyphs

- glyphs
 - can visualize more features of the vector field (flow field)



Arrows and Glyphs

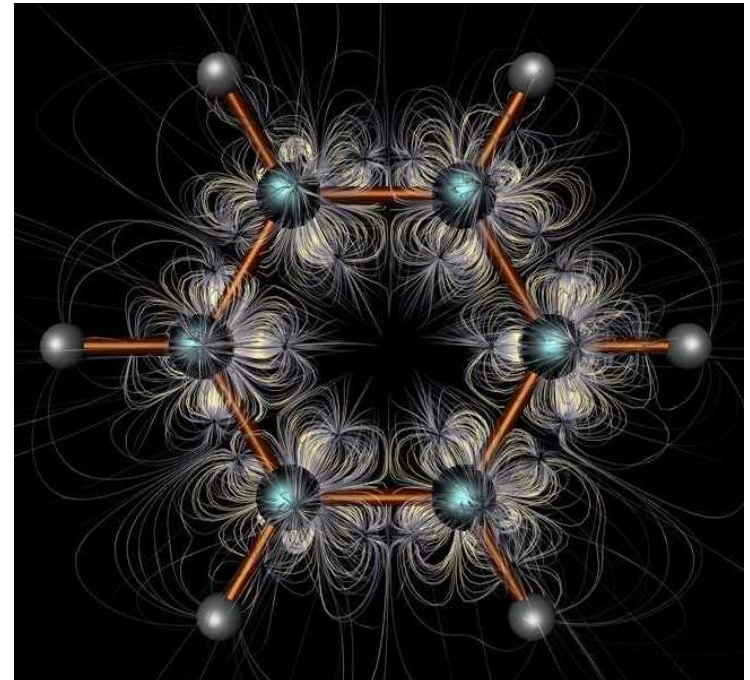
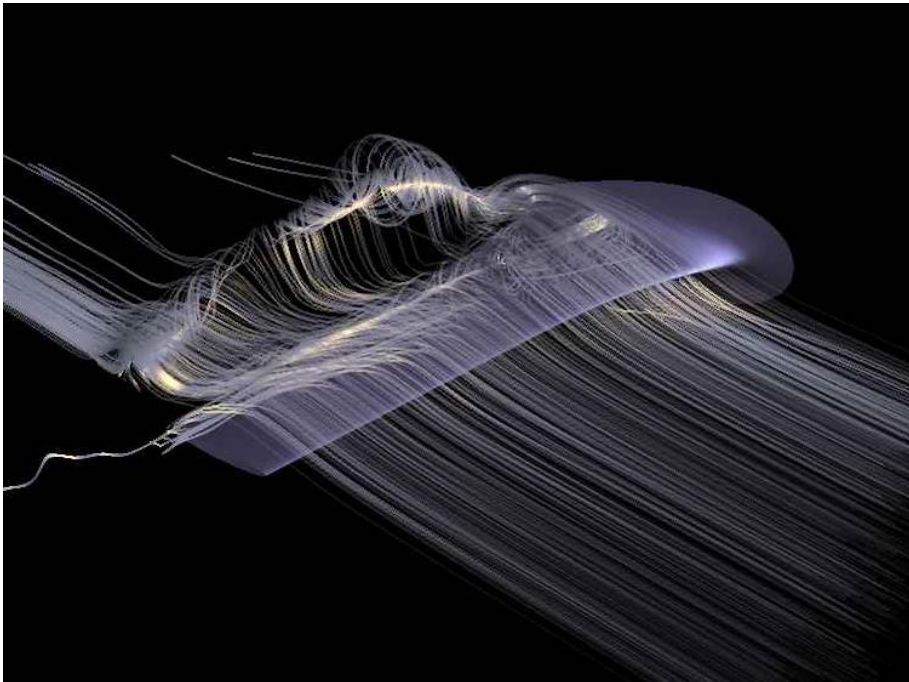
- pros and cons of glyphs and arrows:
 - + simple
 - + 3D effects
 - heavy load in the graphics subsystem
 - inherent occlusion effects
 - poor results if magnitude of velocity changes rapidly (use arrows of constant length and color code magnitude)

Mapping Methods Based on Particle Tracing

- basic idea: trace particles
- characteristic lines
- **local** or **global** methods
- mapping approaches:
 - lines
 - surfaces
 - individual particles
 - sometimes animated

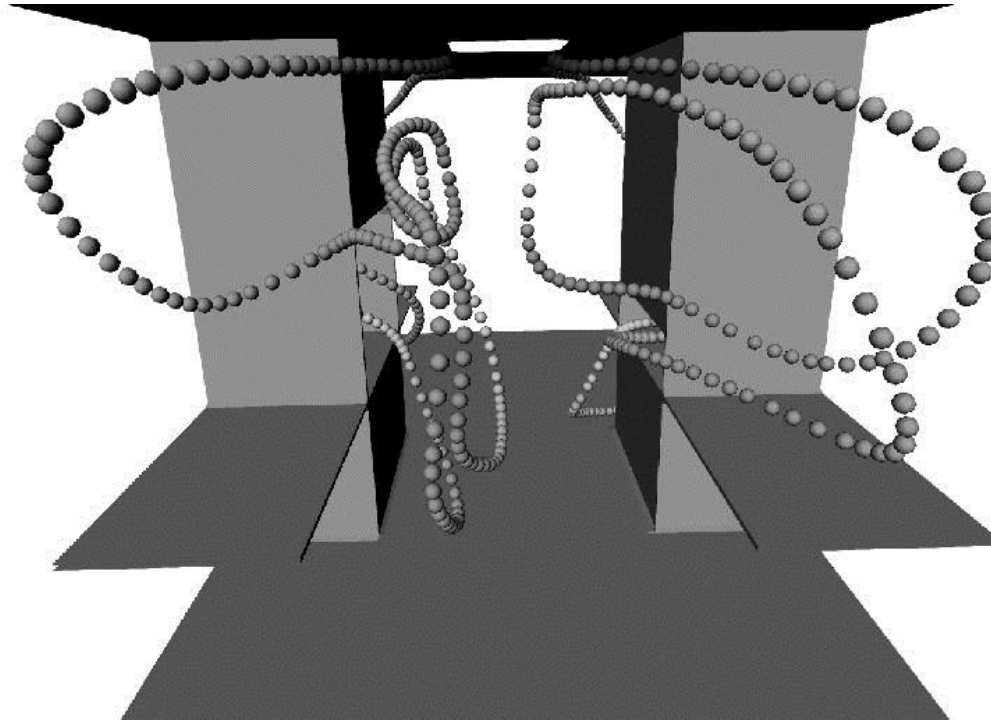
Mapping Methods Based on Particle Tracing

- path lines



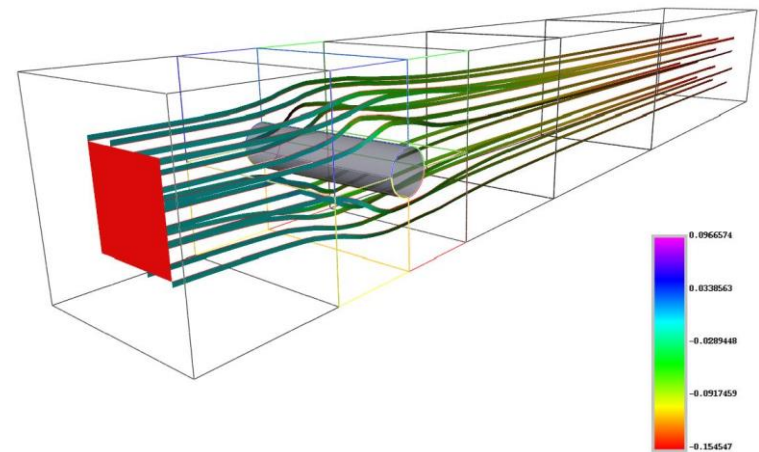
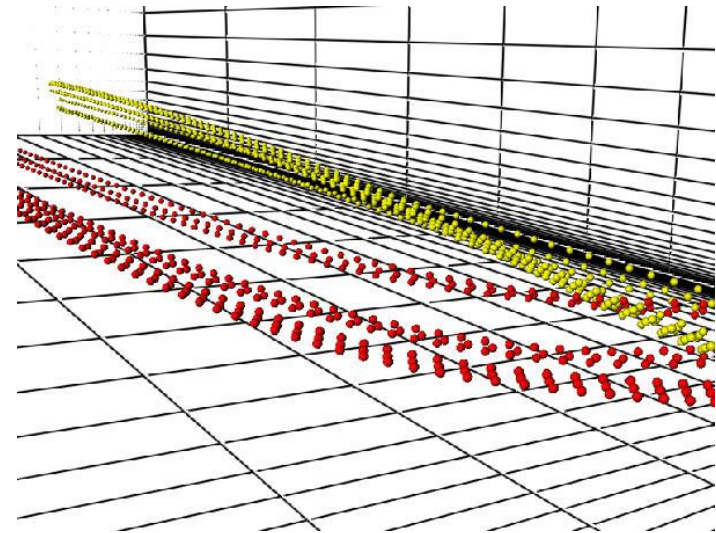
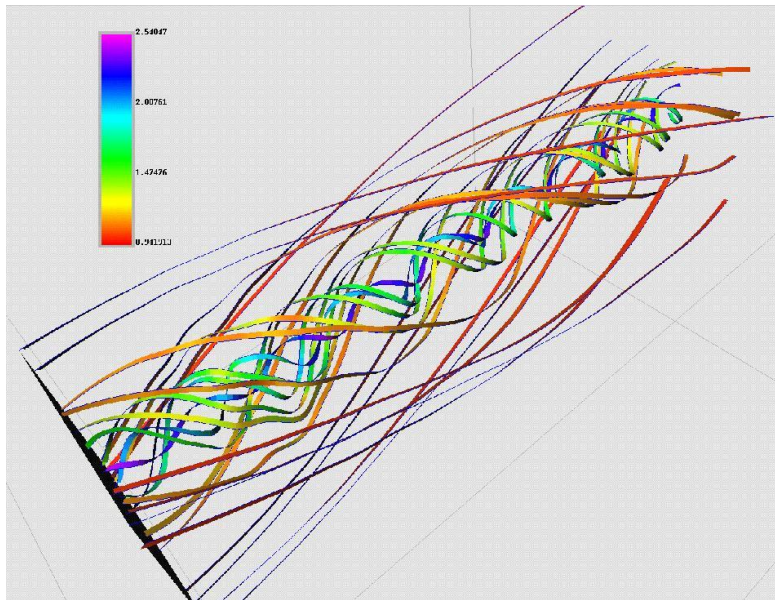
Mapping Methods Based on Particle Tracing

- stream balls
 - encode additional scalar value by radius



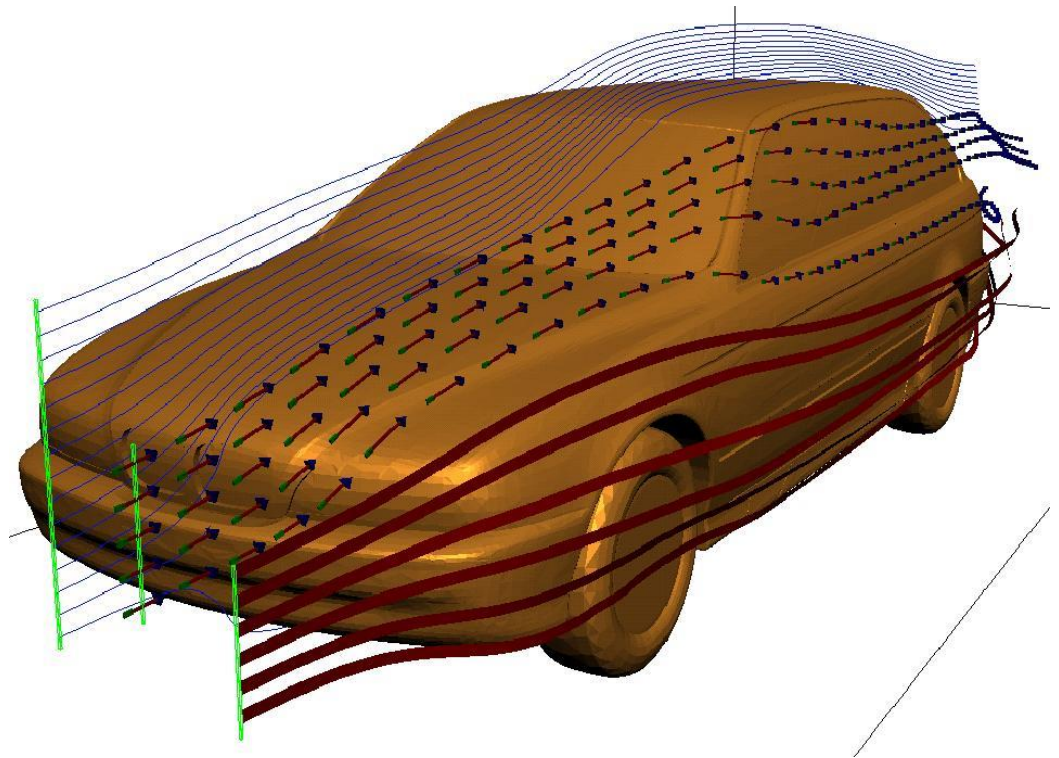
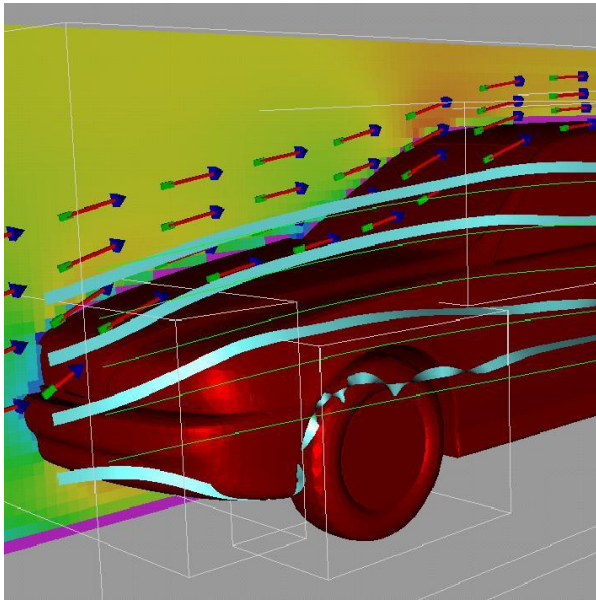
Mapping Methods Based on Particle Tracing

- streak lines



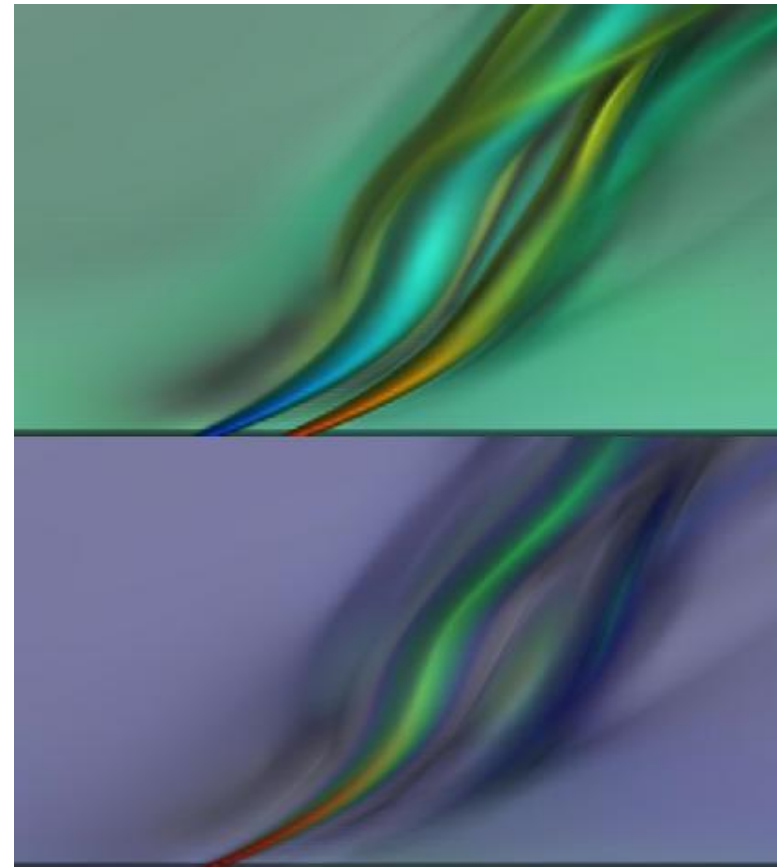
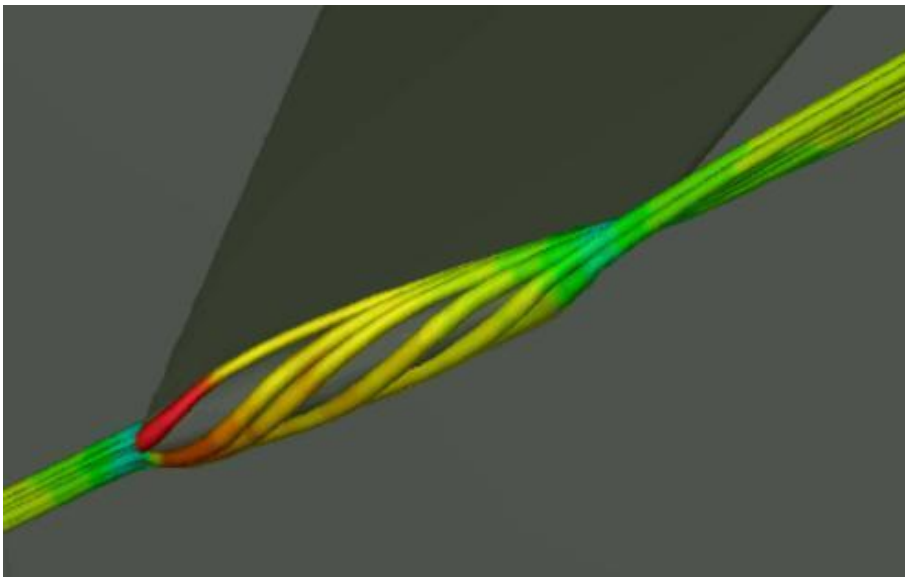
Mapping Methods Based on Particle Tracing

- stream ribbons
 - trace two close-by particles
 - keep distance constant



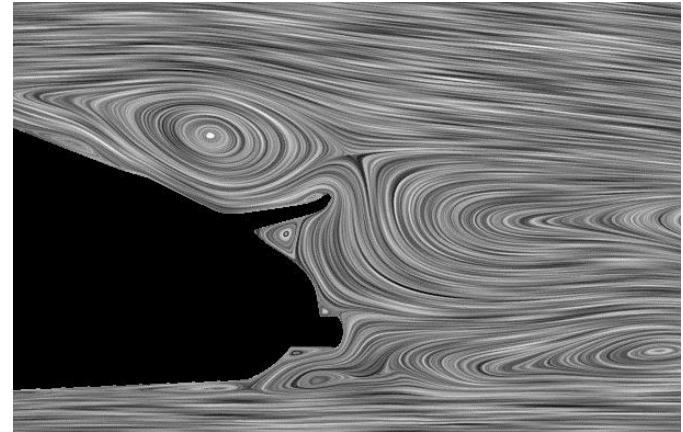
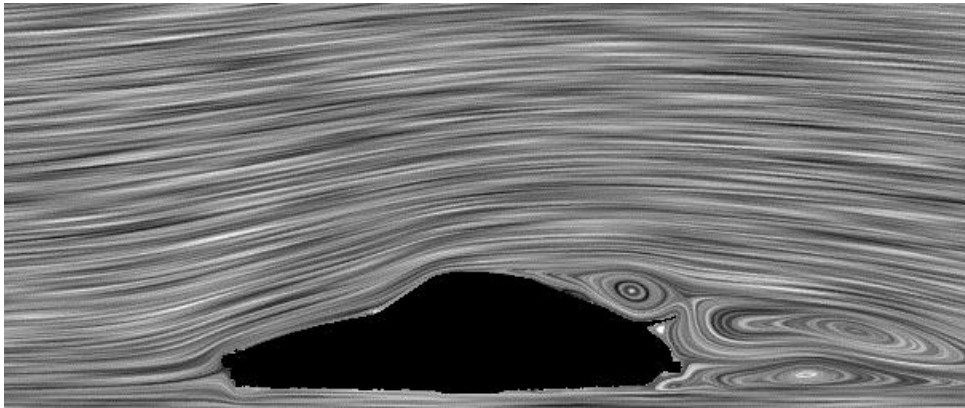
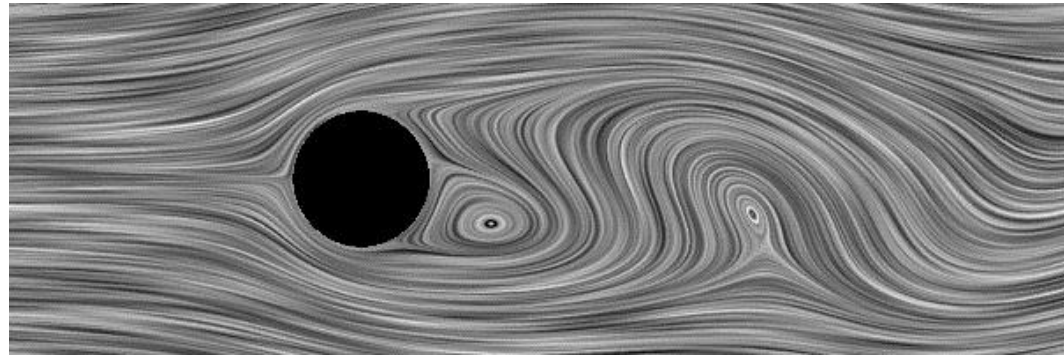
Mapping Methods Based on Particle Tracing

- stream tubes
 - specify contour, e.g. triangle or circle, and trace it through the flow



Mapping Methods Based on Particle Tracing

- LIC (Line Integral Convolution)



Numerical Integration of ODEs

- typical example of particle tracing problem (path line):

$$\mathbf{L}(0) = \mathbf{x}_0 \quad , \quad \frac{d\mathbf{L}(t)}{dt} = \mathbf{v}(\mathbf{L}(t), t)$$

- initial value problem for ordinary differential equations (ODE)
- what kind of numerical solver?

Numerical Integration of ODEs

- rewrite ODE in generic form
- initial value problem for:

$$\dot{\mathbf{x}}(t) = \mathbf{f}(\mathbf{x}, t)$$

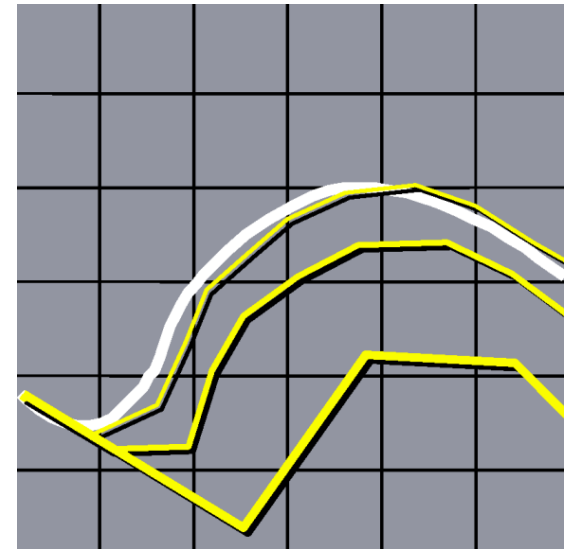
- most simple (naive) approach: explicit Euler method

$$\mathbf{x}(t + \Delta t) = \mathbf{x}(t) + \Delta t \mathbf{f}(\mathbf{x}, t)$$

- based on Taylor expansion

$$\mathbf{x}(t + \Delta t) = \mathbf{x}(t) + \Delta t \dot{\mathbf{x}}(t) + O(\Delta t^2)$$

- first order method
- increase accuracy by smaller step size



Numerical Integration of ODEs

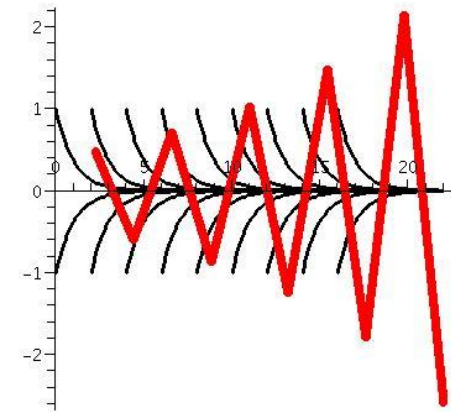
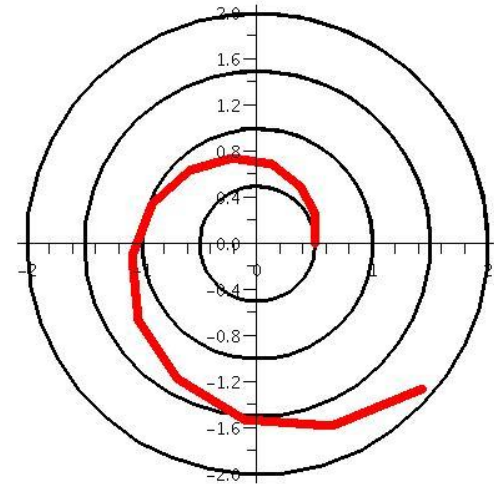
- Problem of explicit Euler method
 - inaccurate
 - unstable

- Example:

$$\dot{x} = -kx$$

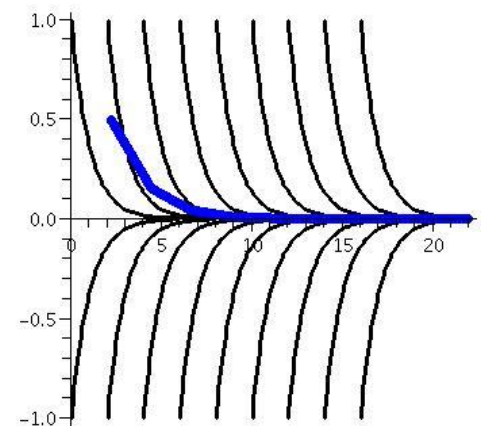
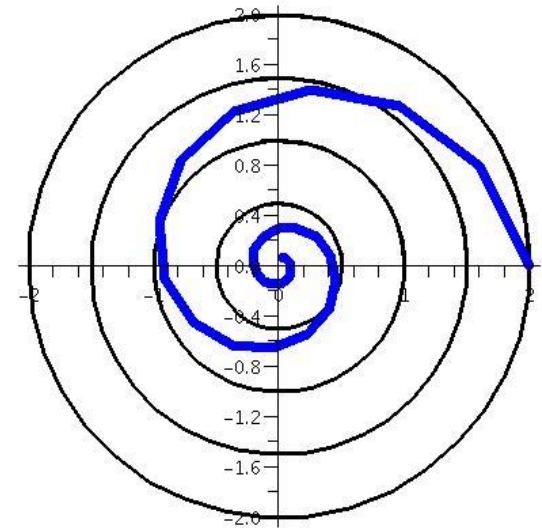
$$x = e^{-kt}$$

divergence for $\Delta t > 2/k$



Numerical Integration of ODEs

- Implicit Euler method
 - Approximation of derivative
$$\dot{\mathbf{x}}(t + \Delta t) \approx \frac{x(t + \Delta t) - x(t)}{\Delta t}$$
 - Implicit timestep
$$x(t + \Delta t) = x(t) + \Delta t \cdot \dot{\mathbf{x}}(t + \Delta t)$$
 - Solution of linear system due to the deriv new position
- Also method of first order



Numerical Integration of ODEs

- Midpoint method

1. explicit Euler step

$$\Delta \mathbf{x} = \Delta t \mathbf{f}(\mathbf{x}, t)$$

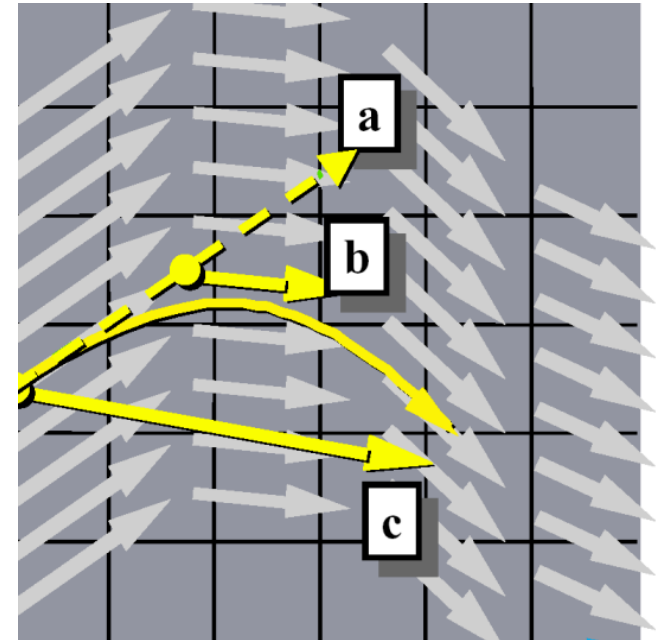
2. Evaluation of \mathbf{f} at midpoint

$$\mathbf{f}_{\text{mid}} = \mathbf{f}\left(\mathbf{x} + \frac{\Delta \mathbf{x}}{2}, t + \frac{\Delta t}{2}\right)$$

3. Complete step with value at midpoint

$$\mathbf{x}(t + \Delta t) = \mathbf{x}(t) + \Delta t \mathbf{f}_{\text{mid}}$$

- Method of second order



Numerical Integration of ODEs

- Classical Runge-Kutta of fourth order

$$\mathbf{k}_1 = \Delta t \mathbf{f}(\mathbf{x}, t)$$

$$\mathbf{k}_2 = \Delta t \mathbf{f}\left(\mathbf{x} + \frac{\mathbf{k}_1}{2}, t + \frac{\Delta t}{2}\right)$$

$$\mathbf{k}_3 = \Delta t \mathbf{f}\left(\mathbf{x} + \frac{\mathbf{k}_2}{2}, t + \frac{\Delta t}{2}\right)$$

$$\mathbf{k}_4 = \Delta t \mathbf{f}(\mathbf{x} + \mathbf{k}_3, t + \Delta t)$$

$$\mathbf{x}(t + \Delta t) = \mathbf{x}(t) + \frac{\mathbf{k}_1}{6} + \frac{\mathbf{k}_2}{3} + \frac{\mathbf{k}_3}{3} + \frac{\mathbf{k}_4}{6} + O(\Delta t^5)$$

Numerical Integration of ODEs

- adaptive stepsize control
 - change step size according to the error
 - decrease/increase step size depending on whether the local error is high/low
 - higher integration speed in “simple” regions
 - good error control
- approaches:
 - stepsize doubling
 - embedded Runge-Kutta schemes
- further reading:
 - WH Press, SA Teukolsky, WT Vetterling, BP Flannery: *Numerical Recipes*

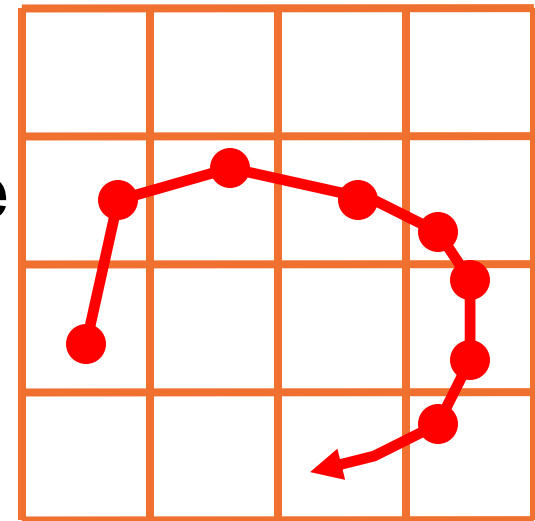
Particle Tracing on Grids

- vector field given on a grid
- solve

$$\mathbf{L}(0) = \mathbf{x}_0 \quad , \quad \frac{d\mathbf{L}(t)}{dt} = \mathbf{v}(\mathbf{L}(t), t)$$

for the path line

- incremental integration
- discretized path of the particle



Particle Tracing on Grids

- most simple case: Cartesian grid
- basic algorithm:

Select start point (seed point)

Find cell that contains start point

While (particle in domain) do

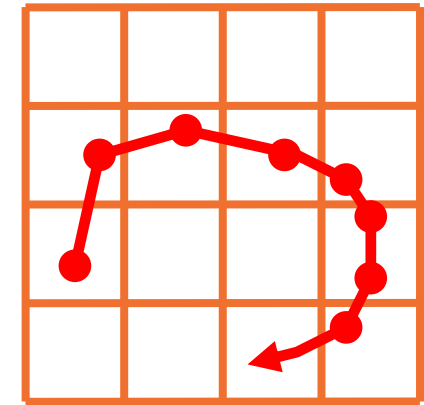
Interpolate vector field at *interpolation*
current position

Integrate to new position *integration*

Find new cell *point location*

Draw line segment between latest
particle positions

Endwhile



point location

interpolation

integration

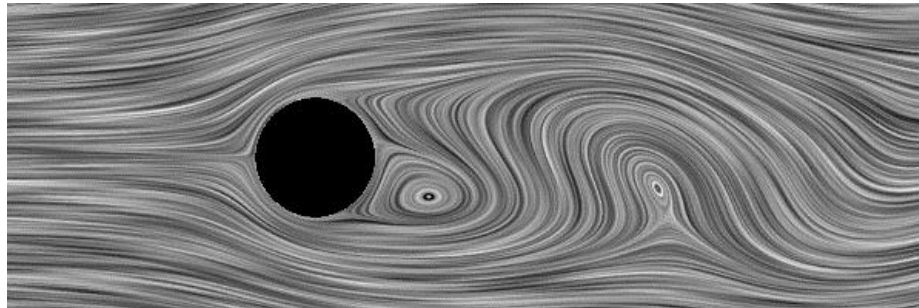
point location

Line Integral Convolution

- a global method to visualize vector fields

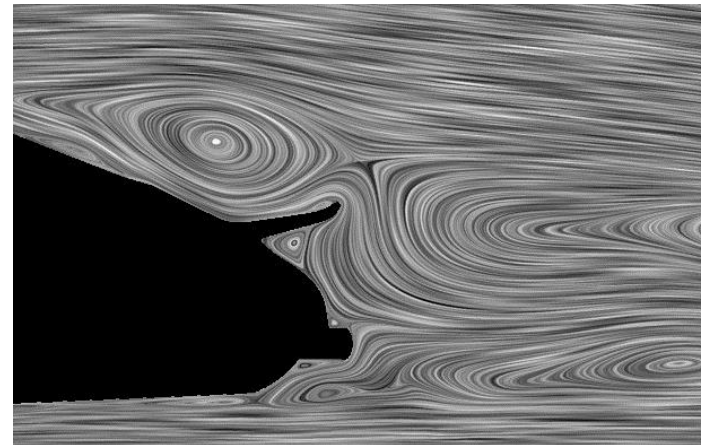
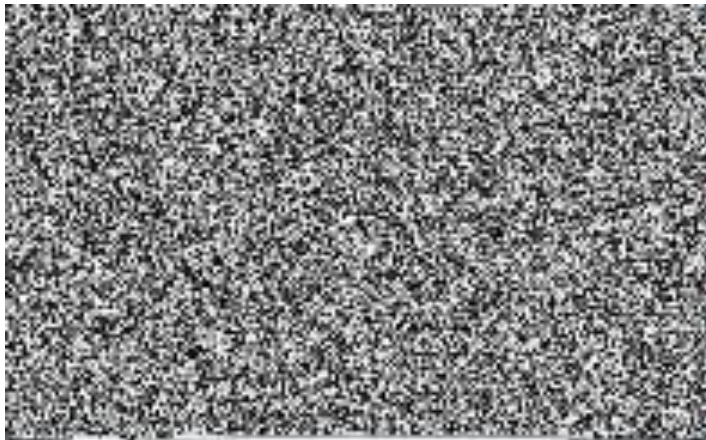
Line Integral Convolution

- line Integral Convolution (LIC)
 - visualize dense flow fields by imaging its integral curves
 - cover domain with a random texture (so called 'input texture', usually stationary white noise)
 - blur (convolve) the input texture along the path lines using a specified filter kernel
- look of 2D LIC images
 - intensity distribution along path lines shows high correlation
 - no correlation between neighboring path lines



Line Integral Convolution

- idea of Line Integral Convolution (LIC)
 - global visualization technique
 - start with random texture
 - smear out along stream lines



Line Integral Convolution

- algorithm for 2D LIC

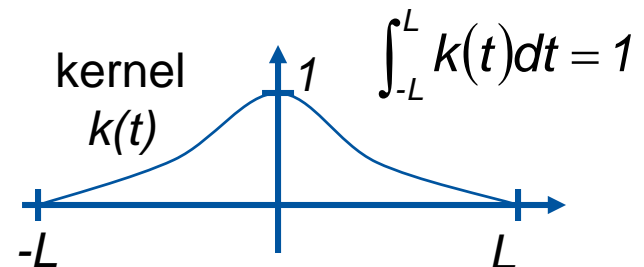
- let $t \rightarrow \Phi_0(t)$ be the path line containing the point (x_0, y_0)
- $T(x, y)$ is the randomly generated input texture
- compute the pixel intensity as:

$$I(x_0, y_0) = \int_{-L}^L k(t) \cdot T(\phi_0(t)) dt$$

convolution with
kernel

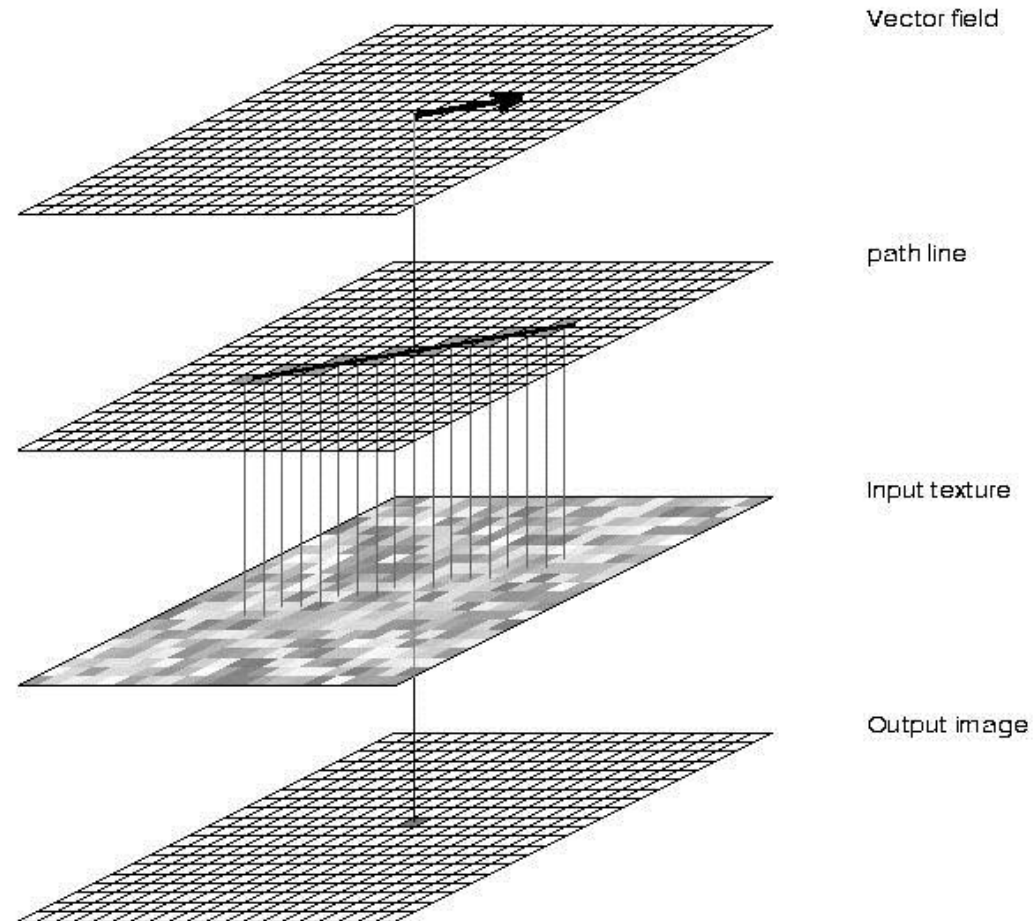
- kernel:

- finite support $[-L, L]$
- normalized
- often simple box filter
- often symmetric (isotropic)



Line Integral Convolution

- algorithm for 2D LIC
 - convolve a random texture along the stream lines



Line Integral Convolution

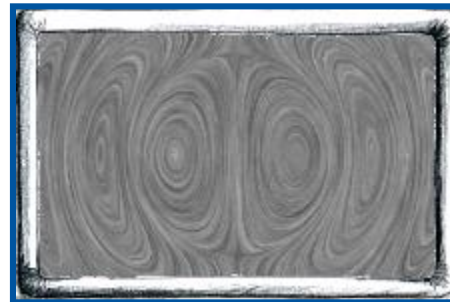
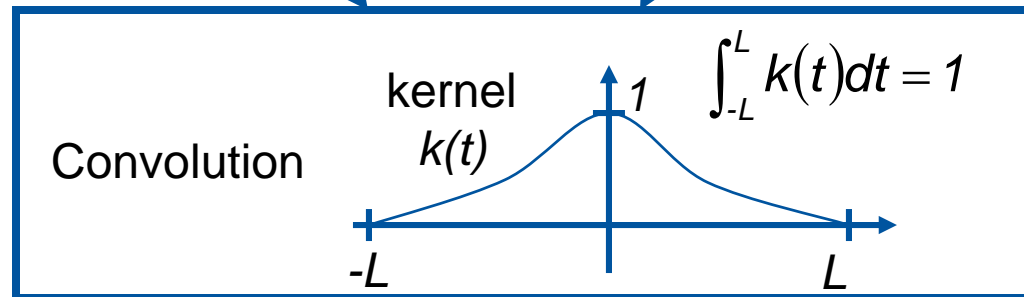


Input noise



Vector Field

Vector field



Final image

Line Integral Convolution

- fast LIC
- problems with LIC
 - new stream line is computed at each pixel
 - convolution (integral) is computed at each pixel
 - slow
- improvement:
 - compute very long stream lines
 - reuse these stream lines for many different pixels
 - incremental computation of the convolution integral

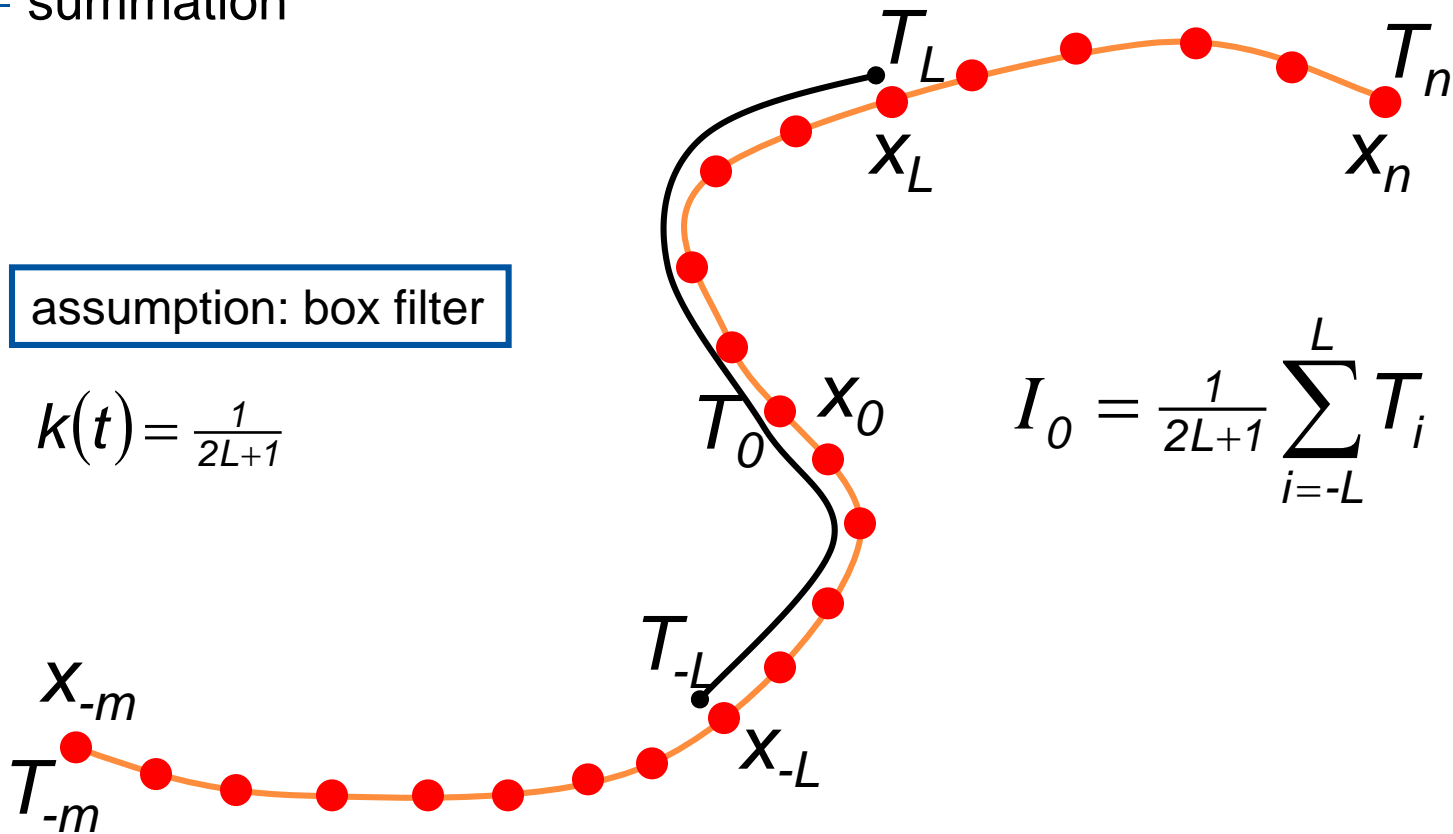
Line Integral Convolution

- fast LIC: incremental integration
 - discretization of convolution integral
 - summation

assumption: box filter

$$k(t) = \frac{1}{2L+1}$$

$$I_0 = \frac{1}{2L+1} \sum_{i=-L}^L T_i$$

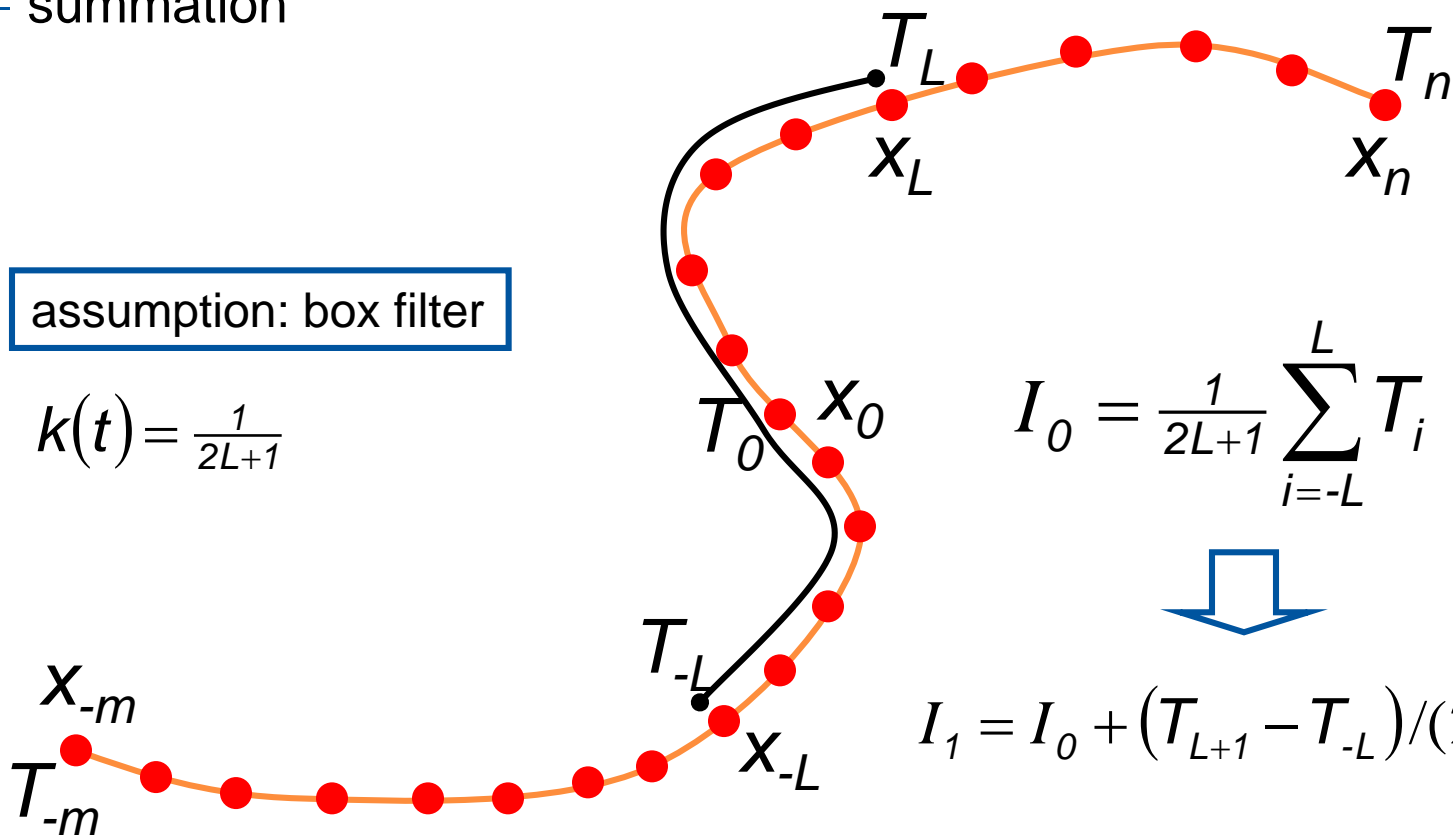


Line Integral Convolution

- fast LIC: incremental integration
 - discretization of convolution integral
 - summation

assumption: box filter

$$k(t) = \frac{1}{2L+1}$$



Line Integral Convolution

- fast LIC: incremental integration for constant kernel
 - stream line $\mathbf{x}_{-m}, \dots, \mathbf{x}_0, \dots, \mathbf{x}_n$ with $m, n \geq L$
 - given texture values $T_{-m}, \dots, T_0, \dots, T_n$
 - what are results of convolution: $I_{-m+L}, \dots, I_0, \dots, I_{n-L}$?
 - for box filter (constant kernel):

$$I_0 = \frac{1}{2L+1} \sum_{i=-L}^L T_i$$

- incremental integration:

$$I_{j+1} - I_j = \frac{1}{2L+1} \sum_{i=-L}^L (T_{i+j+1} - T_{i+j}) = \frac{1}{2L+1} (T_{L+j+1} - T_{-L+j})$$

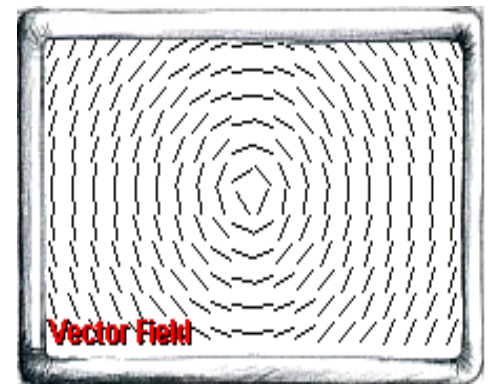
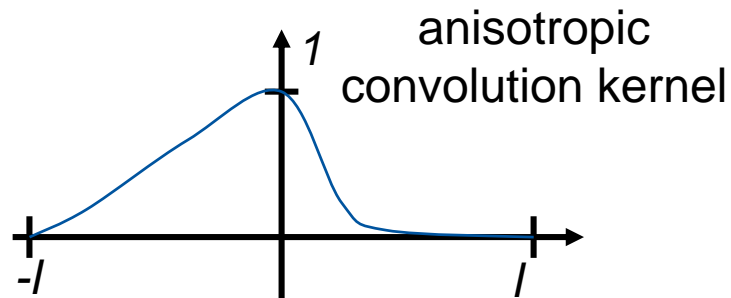
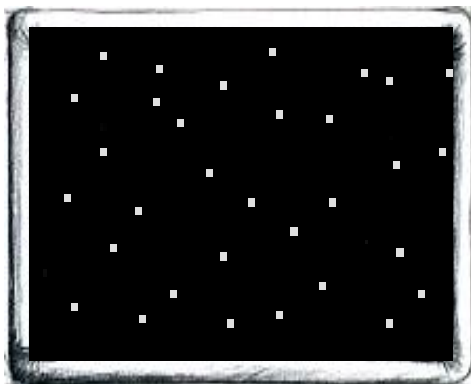
Line Integral Convolution

- fast LIC: Algorithm
 - data structure for output: Luminance/Alpha image
 - luminance = gray-scale output
 - alpha = number of streamline passing through that pixel

```
For each pixel  $p$  in output image
  If (Alpha( $p$ ) < #min) Then
    Initialize streamline computation with  $x_0$  = center of  $p$ 
    Compute convolution  $I(x_0)$ 
    Add result to pixel  $p$ 
    For  $m = 1$  to Limit  $M$ 
      Incremental convolution for  $I(x_m)$  and  $I(x_{-m})$ 
      Add results to pixels containing  $x_m$  and  $x_{-m}$ 
    End for
  End if
End for
Normalize all pixels according to Alpha
```

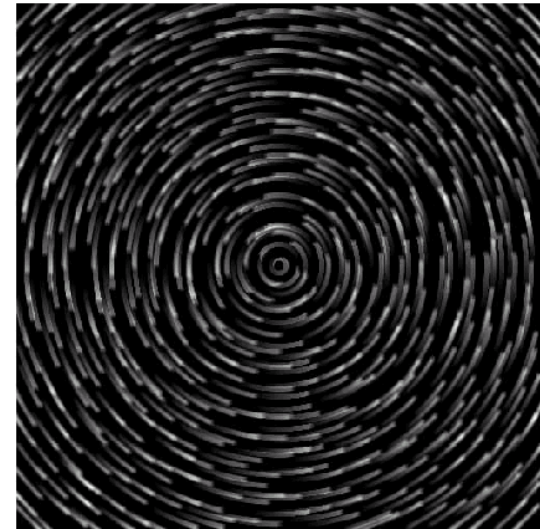
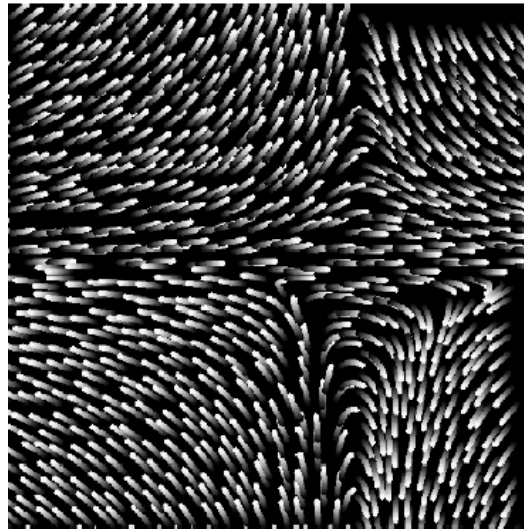

Line Integral Convolution

- oriented LIC (OLIC):
 - visualizes orientation (in addition to direction)
 - sparse texture
 - anisotropic convolution kernel
 - acceleration: integrate individual drops and compose them to final image



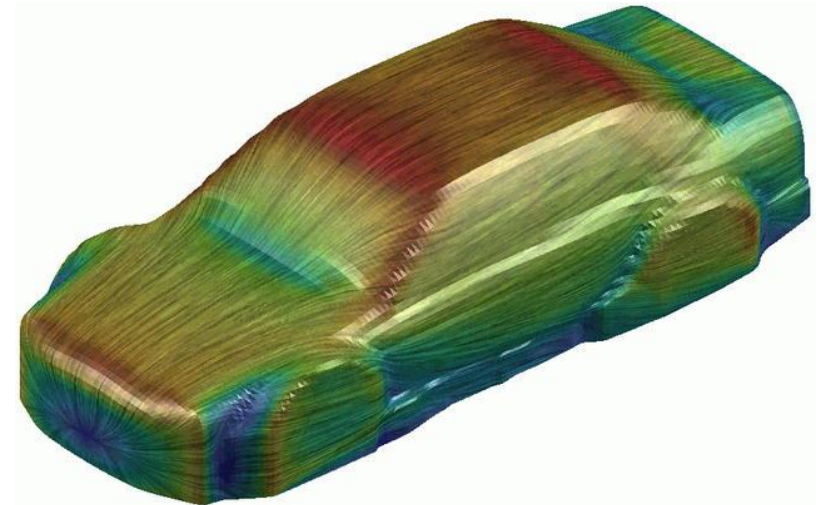
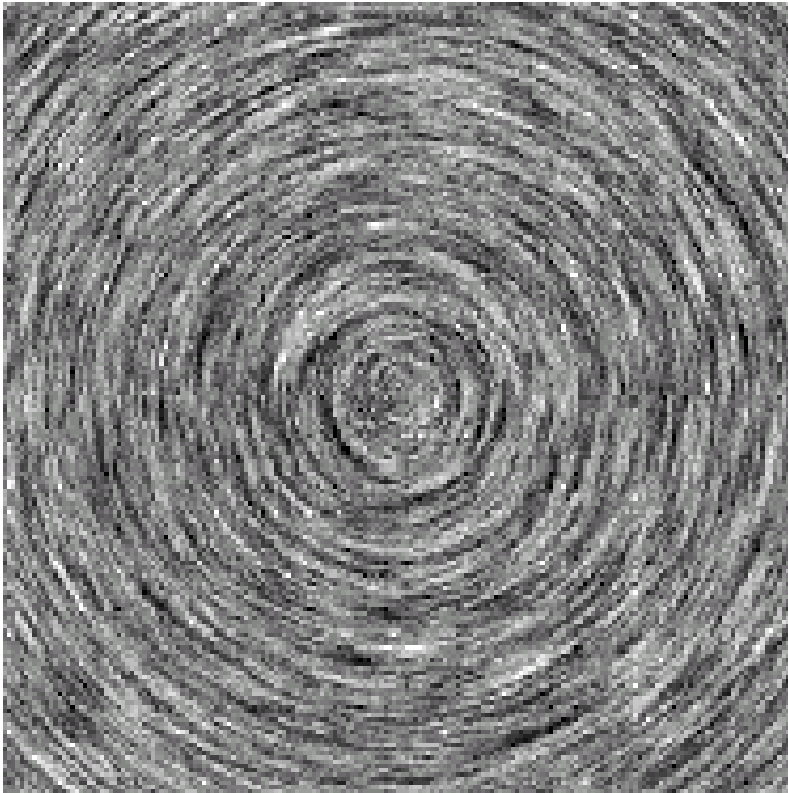
Line Integral Convolution

- oriented LIC (OLIC)

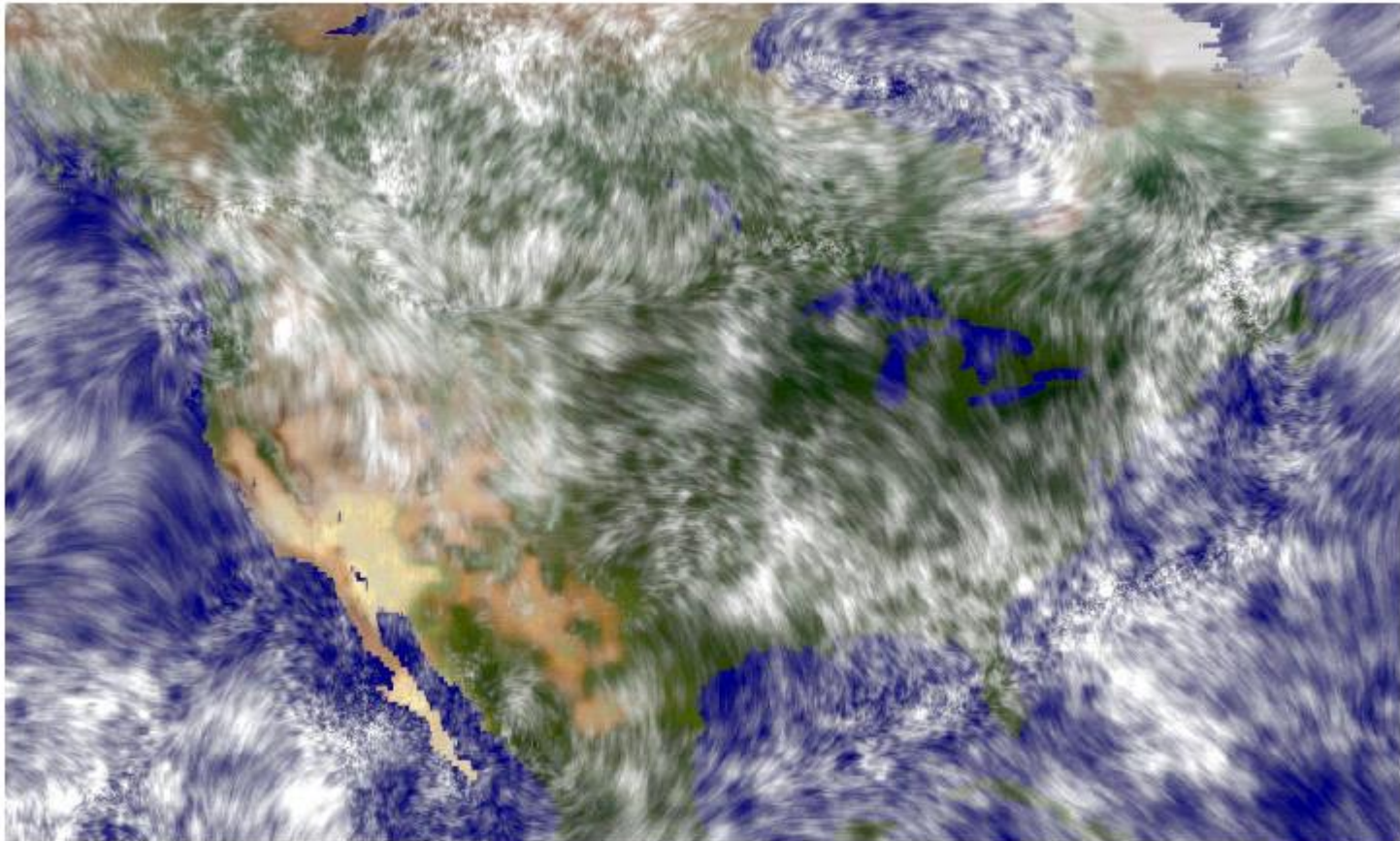


Line Integral Convolution

- LIC - Line Integral Convolution

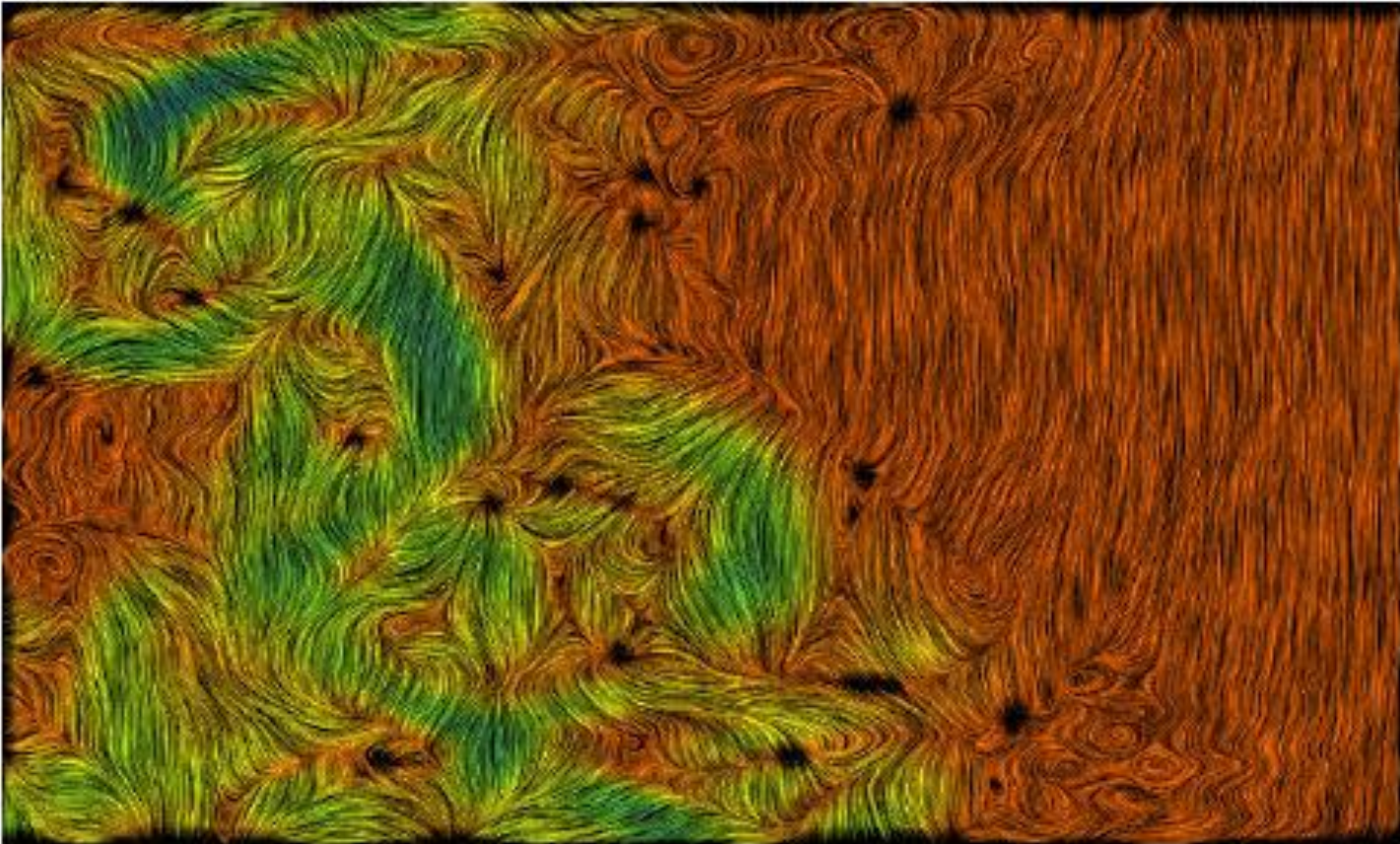


Line Integral Convolution



Lic-Applications: length of convolution integral
with respect to magnitude of vector field

Line Integral Convolution



Lic and color coding of velocity magnitude

Line Integral Convolution

- summary:
 - dense representation of flow fields
 - convolution along stream lines → correlation along stream lines
 - for 2D and 3D flows
 - stationary flows
 - extensions:
 - Unsteady flows
 - Animation
 - Texture advection

3D Vector Fields

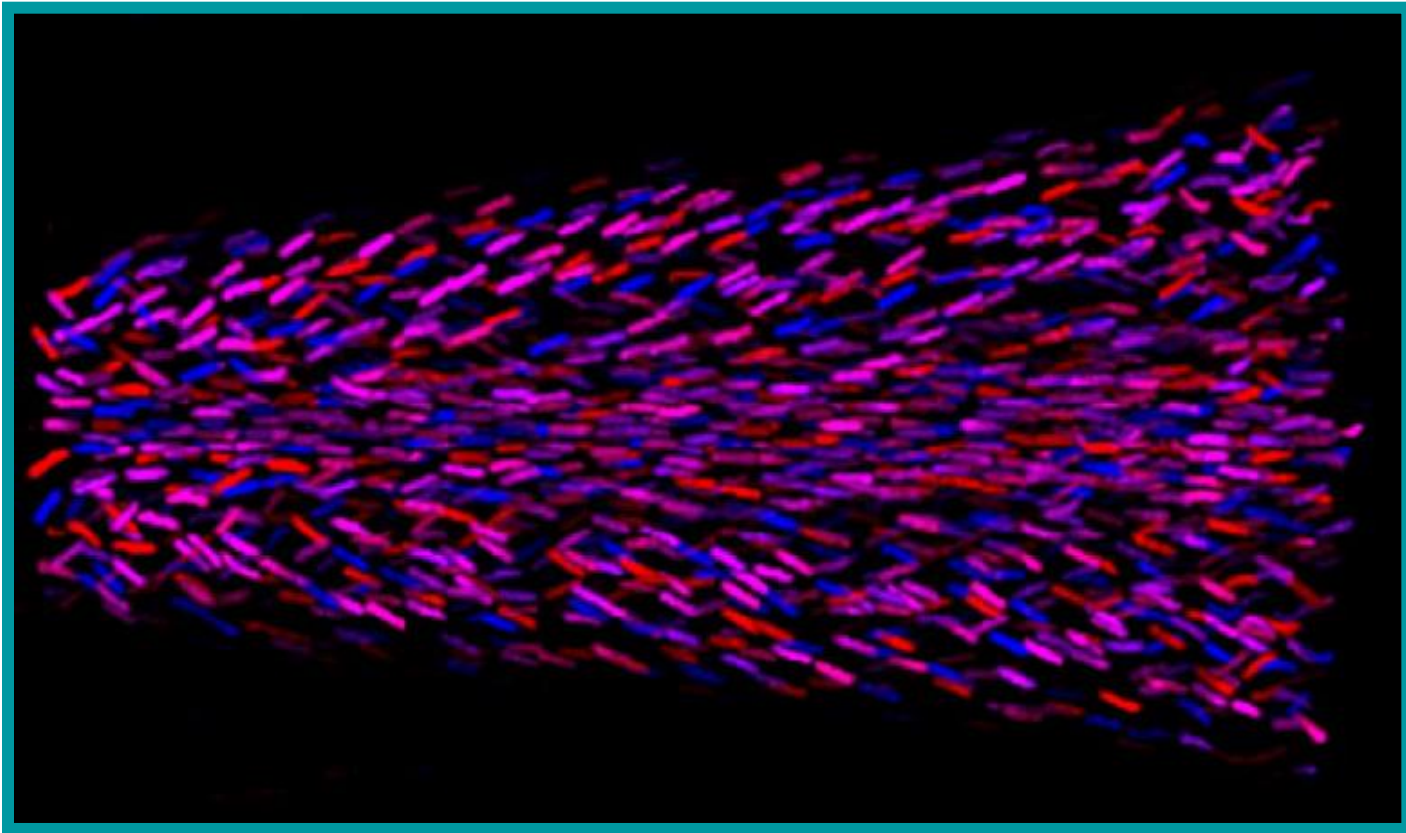
- most algorithms can be applied to 2D and 3D vector fields
- main problem in 3D: effective mapping to graphical primitives
- main aspects:
 - occlusion
 - amount of (visual) data
 - depth perception

3D Vector Fields

- approaches to occlusion issue:
 - sparse representations
 - animation
 - color differences to distinguish separate objects
 - continuity
- reduction of visual data:
 - sparse representations
 - clipping
 - importance of semi-transparency

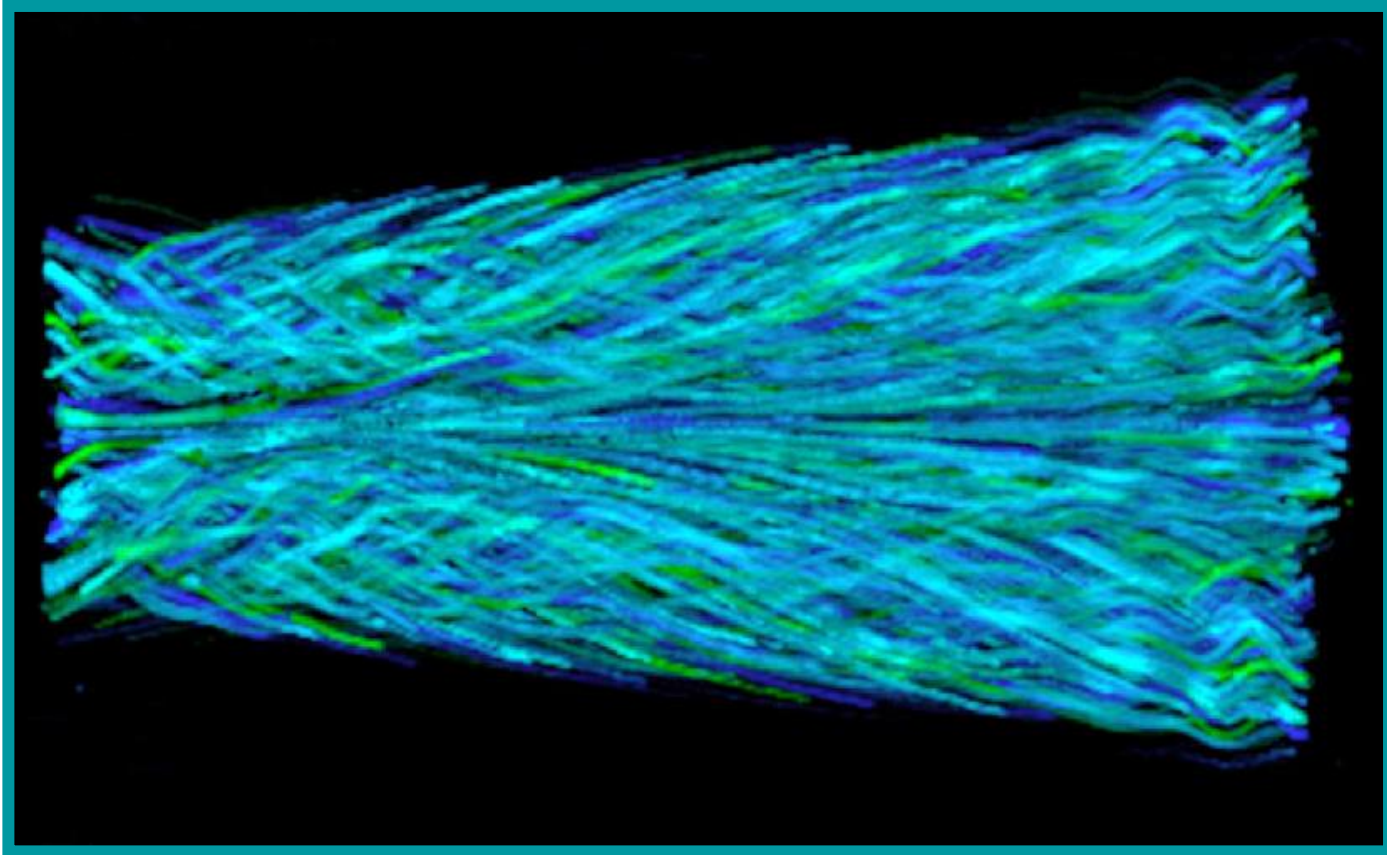
3D Vector Fields

- missing continuity



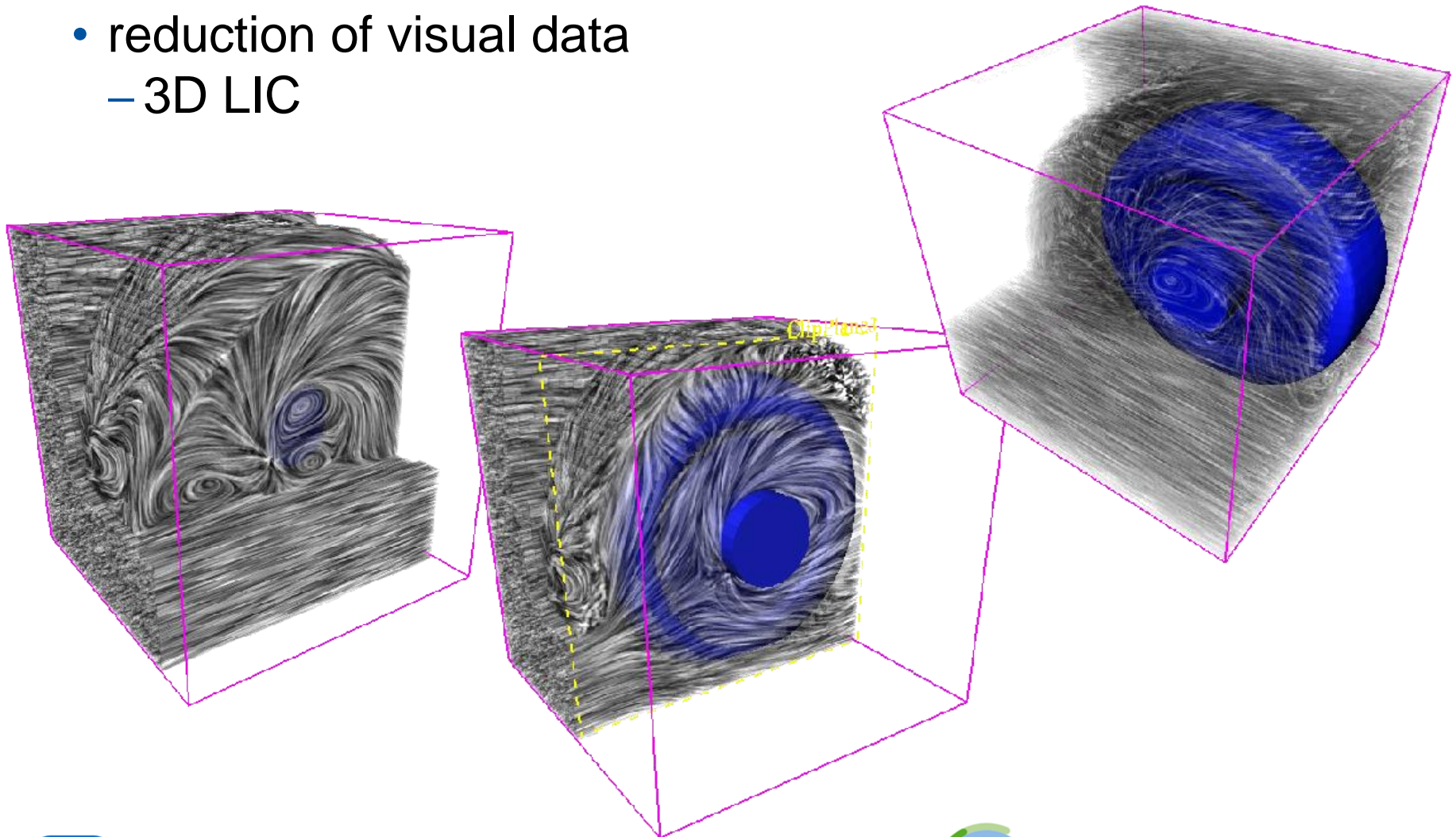
3D Vector Fields

- color differences to identify connected structures



3D Vector Fields

- reduction of visual data
 - 3D LIC

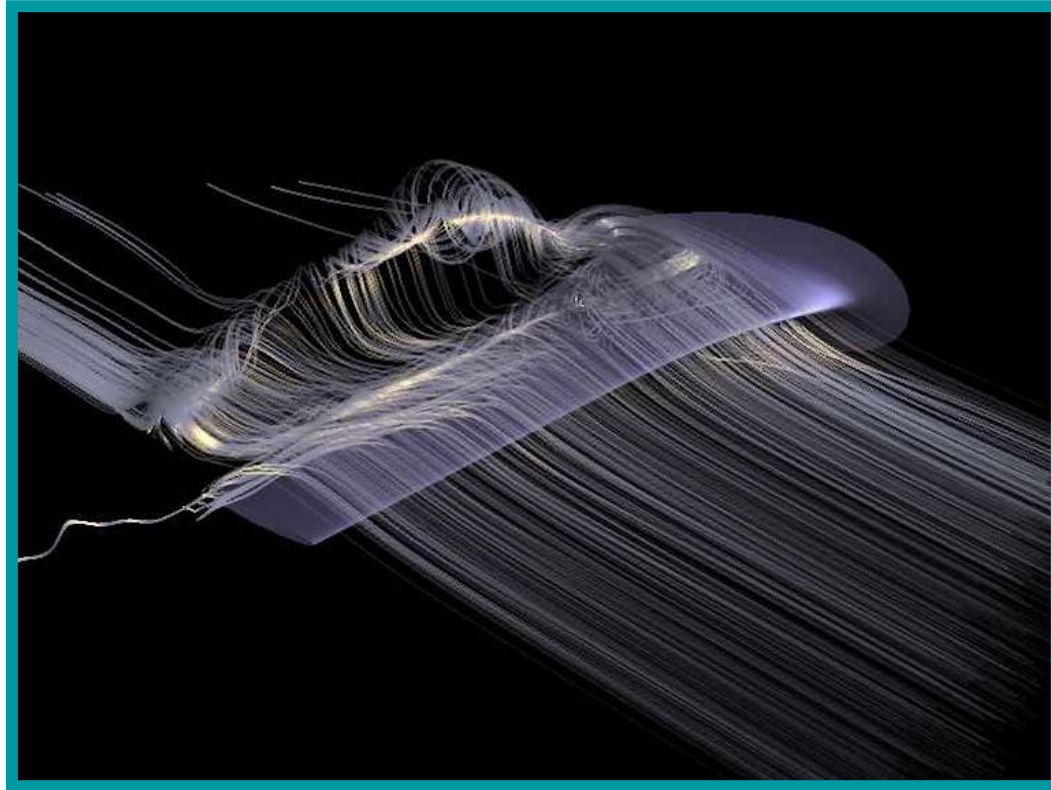


3D Vector Fields

- improving spatial perception:
 - depth cues
 - perspective
 - occlusion
 - motion parallax
 - stereo disparity
 - color (atmospheric, fogging)
 - halos
 - orientation of structures by shading (highlights)

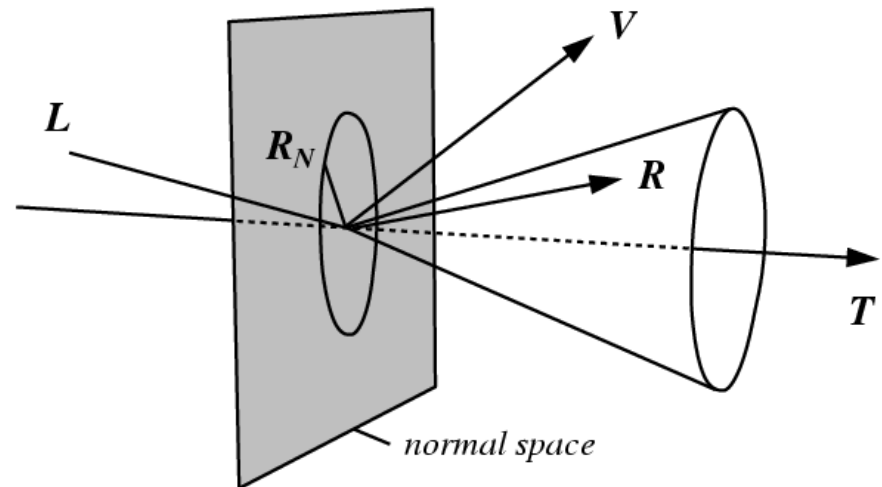
3D Vector Fields

- illumination



3D Vector Fields

- illuminated streamlines [Zöckler et al. 1996]
 - model: streamline is made of thin cylinders
 - problem
 - no distinct normal vector on surface
 - normal vector in plane perpendicular to tangent: normal space
 - cone of reflection vectors



3D Vector Fields

- illuminated streamlines (*cont.*)
 - light vector is split in tangential and normal parts

$$\begin{aligned}\mathbf{V} \cdot \mathbf{R} &= \mathbf{V} \cdot (\mathbf{L}_T - \mathbf{L}_N) = \mathbf{V} \cdot ((\mathbf{L} \cdot \mathbf{T})\mathbf{T} - (\mathbf{L} \cdot \mathbf{N})\mathbf{N}) \\ &= (\mathbf{L} \cdot \mathbf{T})(\mathbf{V} \cdot \mathbf{T}) - (\mathbf{L} \cdot \mathbf{N})(\mathbf{V} \cdot \mathbf{N}) \\ &= (\mathbf{L} \cdot \mathbf{T})(\mathbf{V} \cdot \mathbf{T}) - \sqrt{1 - (\mathbf{L} \cdot \mathbf{T})^2} \sqrt{1 - (\mathbf{V} \cdot \mathbf{T})^2} \\ &= f((\mathbf{L} \cdot \mathbf{T}), (\mathbf{V} \cdot \mathbf{T}))\end{aligned}$$

- Idea: Represent $f()$ by 2D texture
- Access pre-computed $f()$ during rendering

